**Total Questions: 21** 



### Regularization/GLM (Q6L7)

Most Correct Answers: #8 Least Correct Answers: #16

### 1. Overdispersion in Poisson Regression occurs when

**2/10** (A) var(Y|X)>var(Y)

5/10 B var(Y|X)>mean(Y|X)

1/10 C Variance is decreasing

**0/10** (D) I do not know

### 2. Which one of these is the measure for goodness of fit for Poisson Regression?

**0/10** (A) Ordinal R^2

7/10 B Chi-square & Pseudo R^2

**1/10** (c) I do not know

**0/10** D There are not measure for it

# 3. Which one of these is the correct interpretation of the coefficient of Poisson Regression?

1/10 (A) For a 1-unit increase in X, we expect a b1 unit increase in Y.

6/10 B For a 1-unit increase in X, we expect b1 percentage increase in Y.

1/10 C For a 1-percentage increase in X, we expect b1 percentage increase in Y.

0/10 (D) For a 1-percentage increase in X, we expect b1 unit increase in Y.

0/10 (E) I do not know

## 4. In Poisson regression...

3/10 A The asymptotic distribution of the maximum likelihood estimates is multivariate normal.

0/10 (B) The distribution of the maximum likelihood estimates is multivariate normal.

4/10 C The asymptotic distribution of the maximum likelihood estimates is multivariate Poisson distribution.

1/10 D I do not know

A/10  A The likelihood function  4/10  B Chi-squared value  1/10  C I do not know  0/10  D Overdispersion term  6. In the case of intercept-only model  2/10  A The mean of the dependent variable equals the ending of the dependent variable equals the information of the dependent variable equals the information of the dependent variable equals 0  1/10  C The mean of the dependent variable equals 0  1/10  D I do not know  7. In(lambda) = 0.6 - 0.2* female [lamda = the average e^(-0.2)=0.78  3/10  A One unit increase in female brings a 0.2 decrease  3/10  B Being female decreases the average number of an another than the average number	
1/10 © I do not know  0/10 © Overdispersion term  6. In the case of intercept-only model  2/10 A The mean of the dependent variable equals the end  6/10 B The mean of the dependent variable equals the information of the dependent variable equals the information of the dependent variable equals 0  1/10 © The mean of the dependent variable equals 0  1/10 © I do not know  7. In(lambda) = 0.6 - 0.2* female [lamda = the average e^(-0.2)=0.78  3/10 A One unit increase in female brings a 0.2 decrease  3/10 B Being female decreases the average number of an average number of average number of averag	
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<ul> <li>O/10 D I do not know</li> <li>7. In(lambda) = 0.6 - 0.2* female [lamda = the ave e^(-0.2)=0.78</li> <li>3/10 A One unit increase in female brings a 0.2 decrease</li> <li>3/10 B Being female decreases the average number of an average number of average number of an average number of an average number of average number of</li></ul>	itercept
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3/10 Being female decreases the average number of an	erage number of articles] Note:
	in ln(lambda).
3/10 Being female decreases the average number of a	ticles by 0.78 percent
	ticles by 22%
0/10 D I do not know	
8. While running the Poisson Regression we will lambda	have never faced with the value of
<b>7/10</b>	
<b>1/10</b> B 1	
<b>0/10</b>	
1/10 D I do not know	
9. Why does not quasi-Poisson model have AIC?	
5/10 A Quasi-Poisson is used quasi-likelihood instead of	og-likelihood estimates.
3/10 B Quasi-Poisson does not use iterative estimation	
1/10 C I do not know	

10.	why Poisson regression is called log-linear?
5/10	A Because we use a log link to estimate the logarithm of the average value of the dependent variable
1/10	B Because we use a log values of independent variable
1/10	© Because we use a log value of an independent variable is transformed to linear
1/10	D I do not know
11.	In the multiple linear regression, we assume that
4/10	A The number of observations is much larger than the number of variables (n>>p)
2/10	B The number of observations is slightly larger than the number of variables (n>p)
1/10	C The number of observations equals than the number of variables (n=p)
0/10	D The number of observations is lees than the number of variables (n <p)< th=""></p)<>
1/10	E It is not important
0/10	F I do not know
12.	The way of solving the problem of a large number of variables is
3/10	A Subset Selection & Shrinkage (Regularization)
2/10	B Shrinkage (Regularization) & Maximum Likelihood estimation
2/10	© Dimension Reduction & OLS estimation
1/10	D I do not know
1/10	E The absence of the right answer
13.	The bias of an estimator (e.g. z^) equalsHint: the OLS coefficients are unbias :)
3/10	A E(z^) - z
3/10	B E(z^2) - [E(z)]^2
0/10	© [E(z^2) - E(z)]^2
0/10	D E(z^2)
3/10	E I do not know

14.	Which of following is not a type of regularization:
0/10	A L1 - Lasso
0/10	B L2 - Ridge
0/10	© Elastic Net
5/10	D L3 - Passo
4/10	E I do not know
15.	The main idea of regularization is
2/10	To introduce a small amount of bias in order to have less variance.
1/10	B To introduce a small amount of variance in order to have less bias.
2/10	© To introduce a small amount of variance and bias in order to have less bias.
3/10	D I do not know
16.	With which function we can show regularization in R
10.	With Which falletion we can show regularization in K
1/10	A glmnet()
	A glmnet()  (B) regular()
5/10	
	B regular()
5/10 0/10	B regular() C lm()
5/10 0/10 3/10 0/10	B regular() C Im() D glm() E I do not know
5/10 0/10 3/10 0/10 17.	B regular() C lm() D glm() E I do not know  How the tune of any parametr can be made
5/10 0/10 3/10 0/10 17. 3/10	B regular() C lm() D glm() E I do not know  How the tune of any parametr can be made A using Cross validation
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5/10 0/10 3/10 0/10 17. 3/10 1/10	B regular() C Im() D glm() E I do not know  How the tune of any parametr can be made A using Cross validation B It is impossible
5/10 0/10 3/10 0/10 17. 3/10 1/10 2/10	B regular() C Im() D glm() E I do not know  How the tune of any parametr can be made A using Cross validation B It is impossible C I do not now

### 18. **Elastic Net is** the combination of L1 and L2 regularization 5/10 the combination of L2 and L3 regularization 0/10 is independent from other types of refularization 0/10 I do not know 3/10 not a type of regularization 0/10 Regularization is used only for 19. Poisson Regression 1/10 Linear Regression 1/10 Logistic Regression 0/10 any regression 4/10 I do not know 1/10 Regularization can solve the problem of 20. heteroscedasticity 1/10 multicollinearity 2/10 autocorrelation 4/10 I do not know 2/10 Multicollinearity occurs when 21. rank(X)<m (m is the number of explanatory variables)</pre> 3/10 $var(\epsilon) = \sigma^2 I$ 1/10 $E(\epsilon)=0$

0/10

1/10

4/10

cov(εi,εj)=const

I do not know